

# PLS Mean Centering

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## Partial Least Squares (PLS) with Mean Centering

Partial least squares usually uses mean-centered data to compute  $C_{pls}$  using the weights from the SIMPLS algorithm with  $r$  latent variables.

```
X = A_train;
Y = C_train;
X0 = X - mean(X,1);
Y0 = Y - mean(Y,1);
[X_loadings, Y_loadings, X_scores, Y_scores, Weights] = pnnl_simpls(X0, Y0, r);

B_pls = Weights * Y_loadings';

C_pls = (A_unknown - mean(A_train,1)) * B_pls + mean(C_train,1);
```

## Partial Least Squares (PLS) without Mean Centering

The following is PLS without mean centering.

```
X = A_train;
Y = C_train;
[X_loadings, Y_loadings, X_scores, Y_scores, Weights] = pnnl_simpls(X, Y, r);
B_pls = Weights * Y_loadings';
C_pls = A_unknown * B_pls;
```

## Combined algorithm

To make it easier to run with the rest of the tools in the PNNL toolbox, we combined mean-centered and non-mean-centered into one function with meanCentered as an optional argument. When meanCentered is not used as an input, then the default is to compute without mean-centered data.

```
function [C_pls, B_pls] = pnnl_pls(A_train, C_train, A_unknown, r, meanCentered)
if nargin < 5
meanCentered = false;
end

X = A_train;
Y = C_train;
if meanCentered
X0 = X - mean(X,1);
Y0 = Y - mean(Y,1);
else
X0 = X;

```

```

Y0 = Y;
end
[X_loadings,Y_loadings,X_scores,Y_scores,Weights] = pnnl_simpls(X0,Y0,r); %%ok<ASGLU>
B_pls = Weights * Y_loadings';
if meanCentered
C_pls = (A_unknown - mean(A_train,1)) * B_pls + mean(C_train,1);
else
C_pls = A_unknown * B_pls;
end
end

```

## Napalm Data

Load the included napalm data to run the PLS algorithms.

```

clearvars
load pnnl_napalm_data
whos

```

Name	Size	Bytes	Class	Attributes
A_train	20x1713	274080	double	
A_unknown	12x1713	164448	double	
C_train	20x3	480	double	
C_validation	12x3	288	double	
ConcentrationUnits	1x4	8	char	
ConstituentNames	1x3	388	cell	
WavenumberLabel	1x20	40	char	
Wavenumbers	1x1713	13704	double	
ans	1x1	8	double	

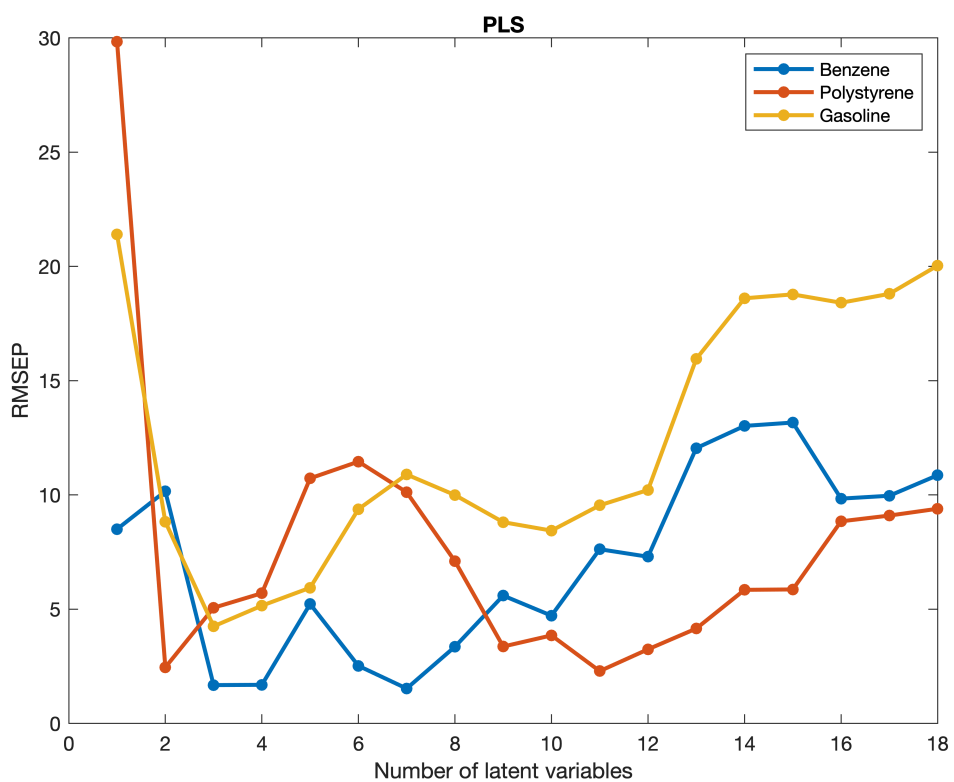
## Optimal number of latent variables for PLS with mean centering

Compute PLS with mean centering for 1 through 18 latent variables and plot RMSEP for them.

```

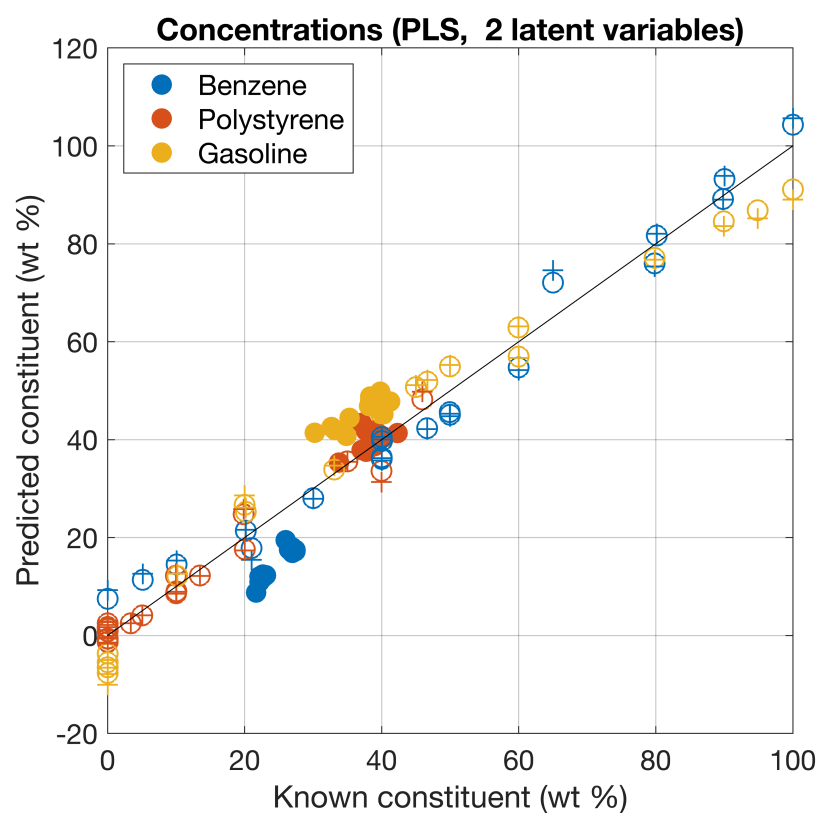
nLatentVariables = 1:18;
meanCentered = true;
[C_pls, RMSEP_pls] = pnnl_napalm_pls(nLatentVariables,meanCentered);
plot(nLatentVariables, RMSEP_pls,'.-','LineWidth',2,'MarkerSize',20)
xlabel('Number of latent variables')
ylabel('RMSEP')
title('PLS')
legend(ConstituentNames{:})

```



For mean-centered data, it looks like the knee in the curve for polystyrene is 2 latent variables, and 3 for benzene and gasoline. Plot them to see what they look like.

```
nLatentVariables = [2,3];  
pnnl_napalm_pls(nLatentVariables,meanCentered);
```

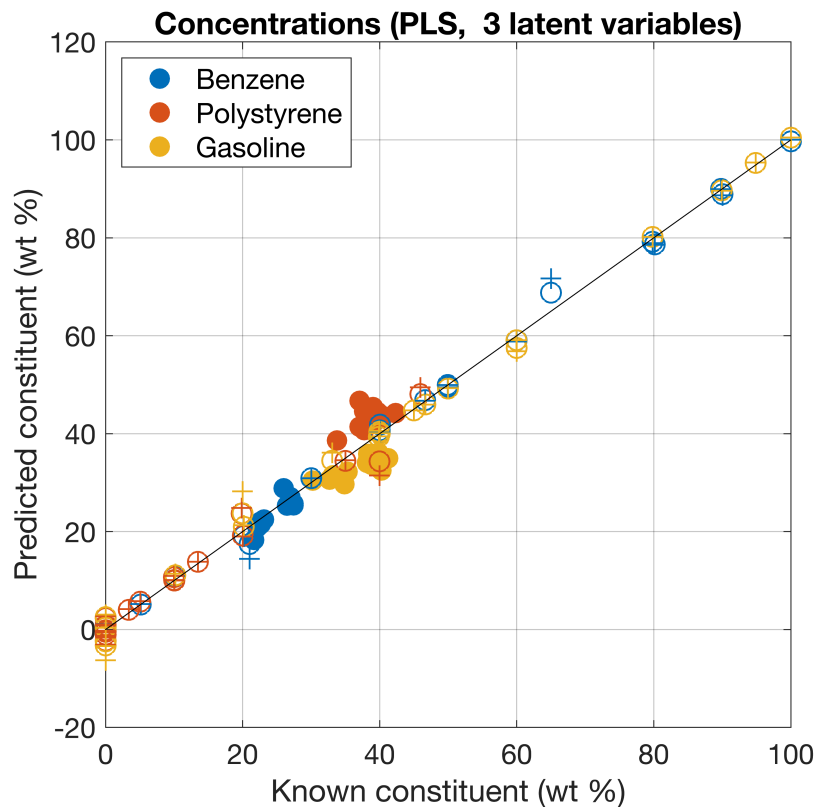


Legend: Dot is predicted. Circle is train. Cross is cross-validation.

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PLS, 2 latent variables	Benzene	Polystyrene	Gasoline
RMSEC	4.1479	2.296	5.5014
RMSECV	5.0229	2.924	6.5258
RMSEP	10.164	2.4566	8.8266

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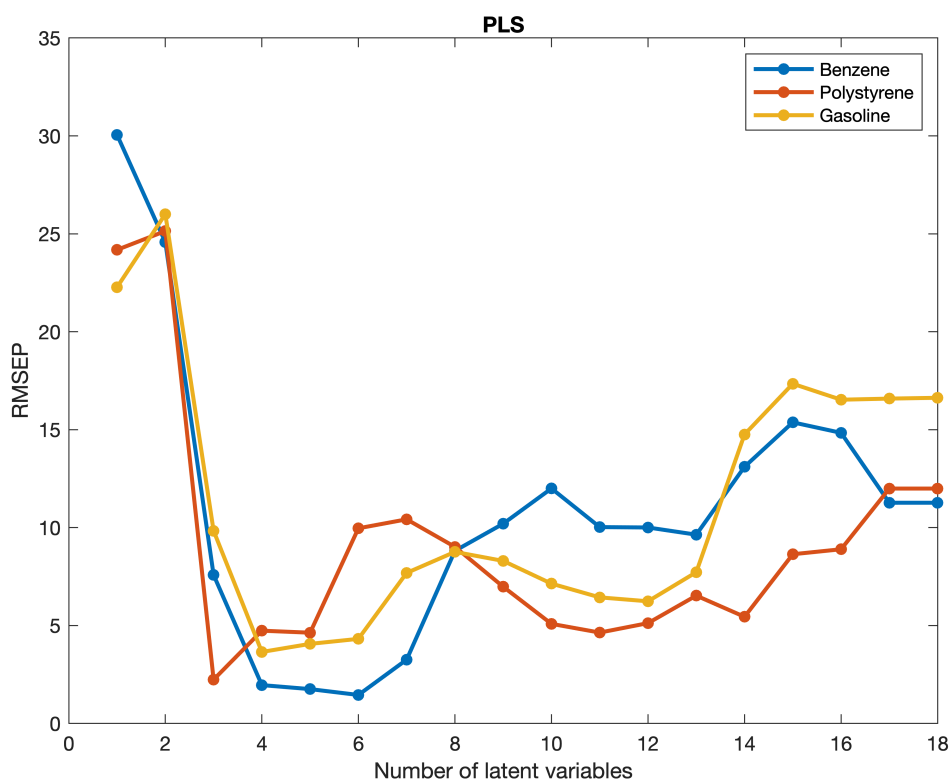
Legend: Dot is predicted. Circle is train. Cross is cross-validation.

PLS, 3 latent variables	Benzene	Polystyrene	Gasoline
RMSEC	1.3968	1.8465	1.5789
RMSECV	2.2406	2.6176	2.7657
RMSEP	1.6799	5.0647	4.2528

## Optimal number of latent variables for PLS without mean centering

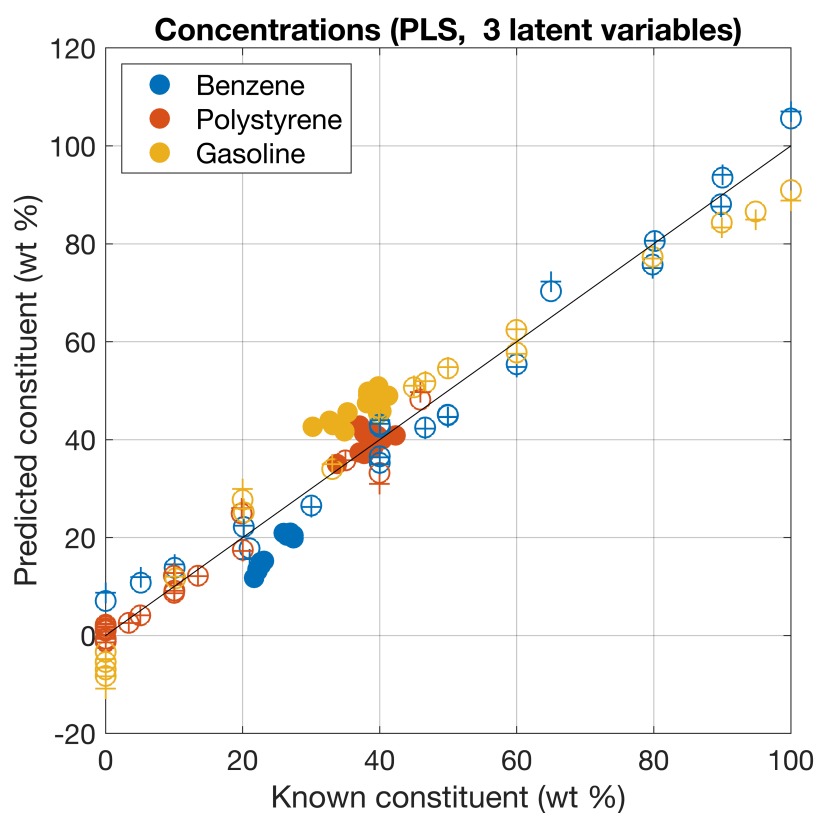
Compute PLS without mean centering for 1 through 18 latent variables and plot RMSEP for them.

```
nLatentVariables = 1:18;
meanCentered = false;
[C_pls, RMSEP_pls] = pnnl_napalm_pls(nLatentVariables,meanCentered);
plot(nLatentVariables, RMSEP_pls,'.-','LineWidth',2,'MarkerSize',20)
xlabel('Number of latent variables')
ylabel('RMSEP')
title('PLS')
legend(ConstituentNames{:})
```



Without mean centering, it looks like the knee in the curve for polystyrene is 3 latent variables, 4 for gasoline, and 6 for benzene. Plot them to see what they look like.

```
nLatentVariables = [3 4 6];  
pnnl_napalm_pls(nLatentVariables,meanCentered);
```

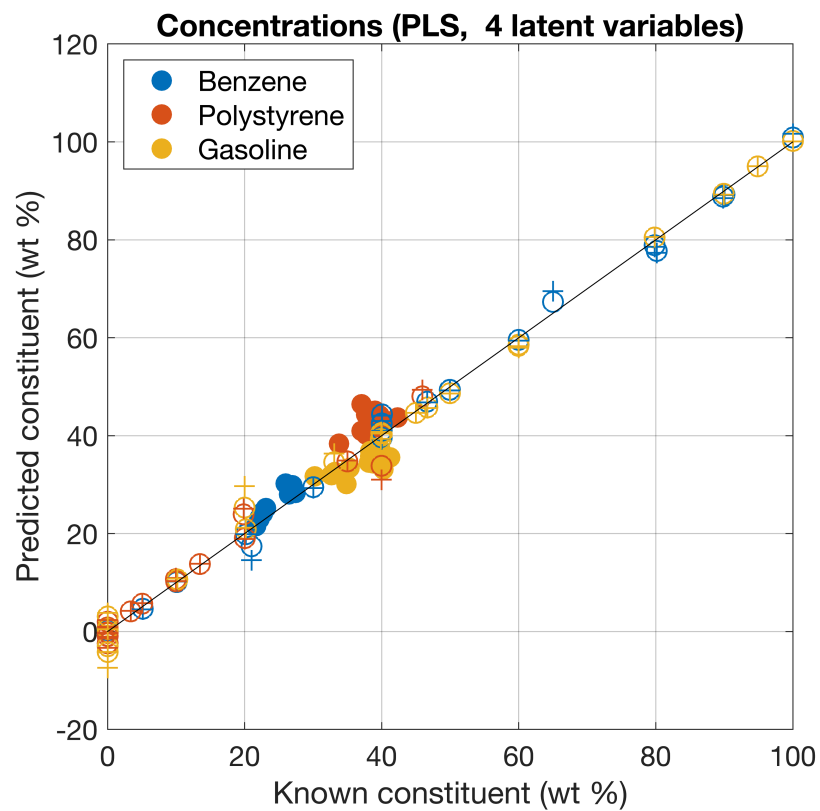


Legend: Dot is predicted. Circle is train. Cross is cross-validation.

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PLS, 3 latent variables	Benzene	Polystyrene	Gasoline
RMSEC	4.1853	2.3592	5.6109
RMSECV	5.0232	2.9897	6.6976
RMSEP	7.5851	2.2404	9.8369

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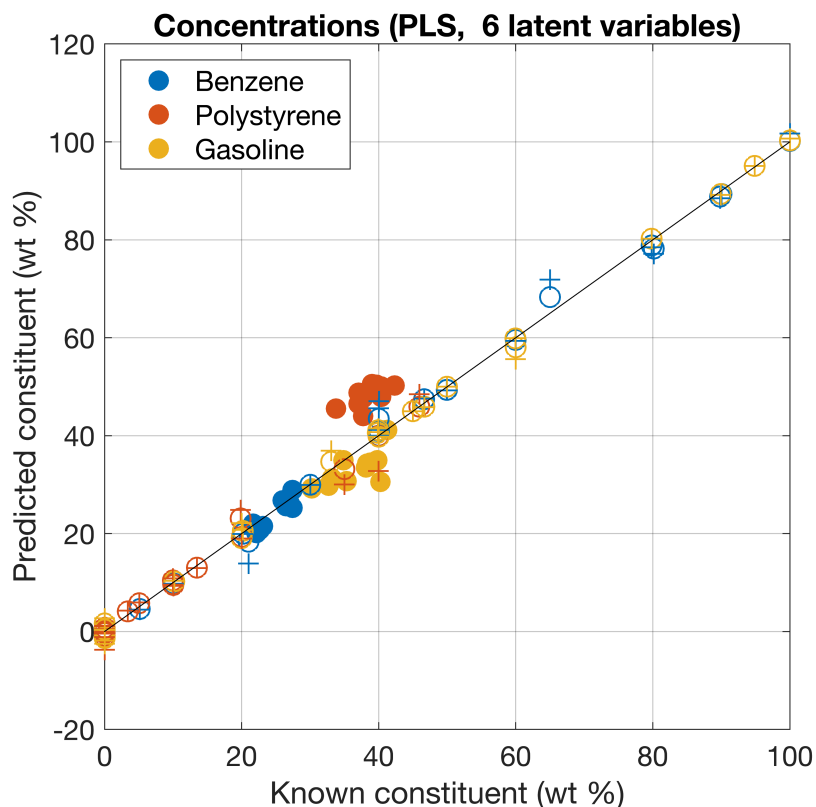
Legend: Dot is predicted. Circle is train. Cross is cross-validation.

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PLS, 4 latent variables	Benzene	Polystyrene	Gasoline
RMSEC	1.6636	1.921	1.9509
RMSECV	2.3112	2.6884	3.2095
RMSEP	1.9599	4.7379	3.6527

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Legend: Dot is predicted. Circle is train. Cross is cross-validation.

PLS, 6 latent variables	Benzene	Polystyrene	Gasoline
RMSEC	1.4134	0.99256	0.91242
RMSECV	3.1544	2.5526	1.7502
RMSEP	1.4567	9.9656	4.3211

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